

#### **Outline**

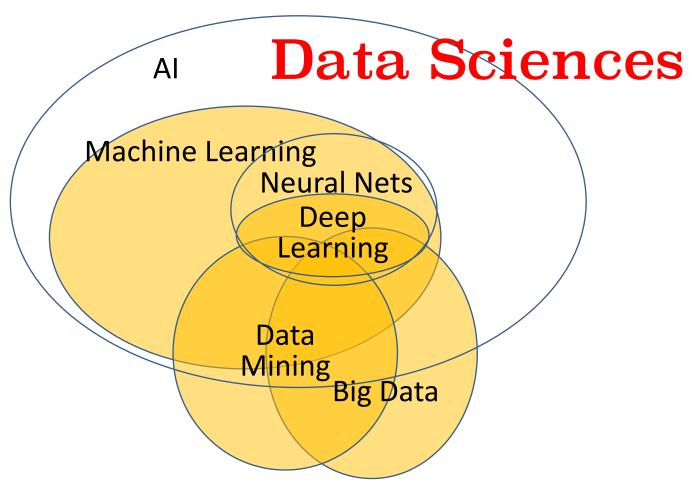


- What Is All This Stuff?
- 2017 Ames Machine Learning Workshop
- Data Sciences Group
- Collaborators



#### What is All This Stuff?

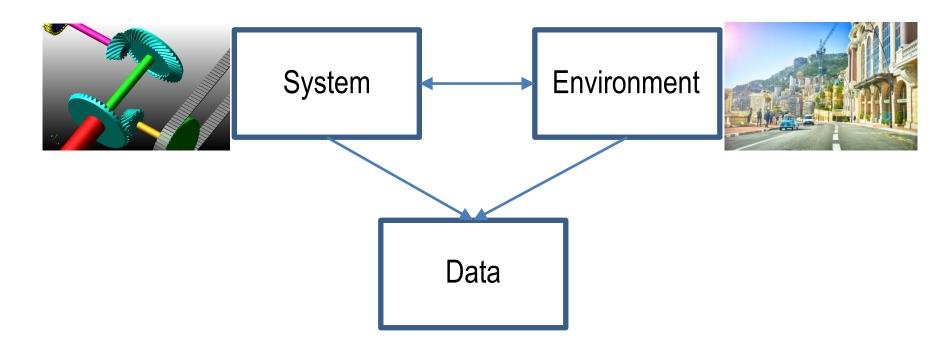






# **Machine Learning**





- Data are produced by system operating in an environment
- Goal: Reverse-engineer system and environment from data
- Understand how system really works, correct system model errors, understand true impact of environment

#### **Outline**



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# 2017 Machine Learning Workshop



- What Machine Learning (ML) has done/can do for NASA problems
- What ML is/is not
  - Is: Difficult, Requires teamwork, Amenable to including domain knowledge, problem info
  - Is Not: Always a big data problem, deep learning + other stuff
- How NASA can engage with academia, industry in advancing ML and domains



# Workshop Agenda



- Keynotes
- Machine Learning for NASA problems
  - Aeronautics
  - Earth Science
  - Space Science
  - Human Space Exploration



# Workshop Agenda



- Keynotes
- Machine Learning for NASA problems
- Technologies Relevant to Machine Learning
  - Human-Machine Interaction
  - Hardware, Program Synthesis, V&V
- Breakout sessions
  - Recommendations for future work
- Path forward



# Keynotes



- Peter Norvig, Director of Research, Google:
  - Practical ML, User provides problem and examples, not how to solve it; more data available---use it
- Vipin Kumar, Professor of Computer Science, University of Minnesota
  - Big Data in Climate, Using physics/Earth science knowledge to guide ML
- Nikunj Oza, Data Sciences Group Leader, NASA Ames Research Center
  - R&D in ML/DM/DS with NASA applications



# Keynotes



- Piyush Mehrotra, HPC Infrastructure for ML
- Mike Little, Advanced Information Systems Technology (AIST), IT for Earth Science



# Machine Learning for Aeronautics



- Deepak Kulkarni: Models of Weather Impact on Airspace Operations
- Heather Arneson: Analysis of Convective Weather Impact on Routing
- Bryan Matthews: Assessing RNAV STAR Adherence
- Vijay Janakiraman: Discovering Precursors to Safety Incidents



#### ML for Aeronautics Breakout



- ML for Safe UAV operations
- Explainable ML for Air Traffic Management Decision Support
- Human Performance Monitoring -> Improved policies, scheduling



## Machine Learning for Space Exploration



- Shawn Wolfe: Automated Monitoring for Mission Operations
- David Thompson: Autonomous Medical Operations
- Rodney Martin: ISHM for Sustainable Habitats
- Adrian Agogino: Machine Learning for Multi-Agent Systems



# ML for Space Exploration Breakout



- Reduce ML application learning curve
- Initially aim for non-safety critical, noncritical path applications to gain trust
- Explainability, V&V critical
- ISHM, assist humans, learn from historical operations



# Machine Learning for Earth Science



- Kamalika Das: ML for Effects of Climate on Amazon Rainforests
- Sangram Ganguly: Deep Learning Models for Satellite Image Classification
- Grey Nearing (Alabama): ML to Improve physics-based hydrology models
- James MacKinnon (NASA GSFC): Deep Neural Nets for Wildfire Detection, offline and embedded
- Stefano Ermon (Stanford): Transfer Learning, Deep Learning for poverty prediction, crop yield prediction

#### ML for Earth Science



- Data: Noisy, multiple spatiotemporal resolutions
- Problems: Mapping (e.g., fire, poverty),
   Prediction (e.g., fires, extreme weather,
   climate), learning from physics and data
- Distributed sensing, intelligent instruments



# Machine Learning for Astrophysics/Planetary



- Hamed Valizadegan: ML for Space Science and Engineering
- Nick Kern (Berkeley): Surrogate Modeling for Cosmology
- Sean McGregor (Oregon): FDL—Deep Learning for Solar Storm Prediction
- Mark Cheung (LockMart): FDL---Data Science for Heliophysics
- Madhulika Guhathakurta (NASA HQ)---FDL Overview

# Astrophysics/Planetary Breakout



- Vetting Transiting Exoplanet Candidates:
   Classification, identifying relevant features
- Better Data Archives: Easier to use and access, facilitate large studies
- Frontier Development Labs (FDL)-type intensive collaborations



# Hardware, Program Synthesis, V&V



- Cliff Young (Google): Tensorflow Processing Unit
- Timothy Randles (LANL): CharlieCloud containers
- Natalia Vassilieva (HP): Hardware and software choices for deep learning
- Johann Schumann: Toward synthesizing ML algorithms
- Guy Katz (Stanford): V&V of deep nets



# Hardware, etc., Breakout



- Accounting for environment: low network bandwidth, onboard computing
- Data: format, too much and too little data
- Air Traffic Control: REALLY understand it, use all relevant data
- ML algorithms: When to use which ones?
- ML for Security, Security for ML
- ML for NASA programs
- Software for ML, including V&V



#### **Human-Machine Interaction**



- Milind Tambe (USC): Human-Machine Partnership for Social Good
- Karen Myers (SRI): Learning to Help Human Problem Solvers
- Kamalika Das: Active Learning for Domain Expert Feedback for Anomaly Detection
- Alonso Vera: What Machines Need to Learn to Help Human Problem Solvers



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## The Data Sciences Group at NASA Ames



Data Mining Research and Development (R&D) for application to NASA problems (Aeronautics, Earth Science, Space Exploration, Space Science)

#### **Group Members**

# Ilya Avrekh Kamalika Das, Ph.D. Dave Iverson Rodney Martin, Ph.D. Bryan Matthews Nikunj Oza, Ph.D. John Stutz Hamed Valizadegan, Ph.D. + students

#### **Funding Sources**

- Science Mission Directorate: AIST and CMAC programs
- NASA Aeronautics Research Mission Directorate- ATD, SMART-NAS, SASO Project, Seedling Fund
- NASA Engineering and Safety Center
- Ames Center Innovation Fund
- AMMOS
- Non-NASA: DARPA, DoD



#### Collaborators



- Universities: Basic research in data sciences, domains
- Industry: Data sources, baseline methods, domain expertise
- NASA: Apply basic research, develop for NASA's needs, funding programs
- Other government: funding, domain expertise, data sources





























































#### **DASHlink**

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DASHlink is a collaborative website designed to promote:

- Sustainability
- Reproducibility
- Dissemination
- Community building

Users can create profiles

- Share papers, upload and download open source algorithms
- Find NASA data sets.







Data Sciences at NASA Ames
2017 Ames Wachine Learning
Workshop
https://ti.arc.nasa.gov/events/
machinelearningworkshop2017/

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